#### Visualization / Data

- Data Participants:
- A. Choudhary
- T. Critchlow
- L. Diachen
- P. Heermann
- R. Moore
- B. Parvin
- N. Samatova
- A. Shoshani
- M. Vouk

- Vis Participants
- A. Breckenridge
- W. Bethel
- P. Crossno
- J. Kohl
- A. McPherson
- M. Papka
- J. van Rosendale

## Petascale Data Implications

- Visualization is a knowledge generation process that requires a human in the loop
- Visualization is a form of qualitative data analysis
- Information extraction(feature detection) is a quantitative process
- Knowledge is the generation of relationships between features
- For petascale data, knowledge generation is essential

#### Petascale Computers

- Petascale computer may be a heterogeneous environment, with varying data locality
  - View petascale computer as a grid linking heterogeneous resources
  - Requires co-scheduling of CPU, data movement, I/O bandwidth
  - Challenge is maintaining parallelism across a hierarchical data management process

#### Barriers

- More data than a person can look at
  - Co-scheduling of CPU, storage, and I/O transport
  - Data management / navigation
  - Scalable algorithms for data analysis
- Information extraction
  - Augment human for generation of knowledge
  - On-the-fly data analysis
- Knowledge generation
- Logical data model abstraction drives knowledge abstraction

## Dealing with Petascale Data

- Petascale data management
- Post processing of data
- Knowledge extraction and characterization

## 1. Petascale Data Management

- Distributed data ingestion
  - Problem set up
  - Fast I/O transport for data intensive computing
  - Storage management co-scheduling with CPU
  - Data pre-staging management / scheduling
- Collection based data management beyond file systems
  - Use of attributes to specify semantic meaning and features
- Knowledge based data management
  - Use of relationships to describe interactions between features in data and semantics of attributes

## 2. Post Processing of Data

- Not just getting petabytes to the desk top
  But also getting information to the desk top
- Data analysis on distributed data
- Levels of abstraction for visualizing data and knowledge
- Integrated compute / analysis / storage
  - Infinite compute with no storage
  - Infinite storage (sensor data)
  - Optimization of storage and compute resource allocation

# 2. Processing Infrastructure

- Data support
  - Simulation data
  - Sensor data
  - Derived data products
- Agent-based data analysis
  - Mediators for data extraction
- Knowledge generation
  - Definition of feature space
  - Visualization of feature space
  - Differentiation between anomalies and implied knowledge

# 2. Grid Technologies

- Dealing with heterogeneity
- Data management Identification of data, navigation through data, organization of data
  - Generation of metadata
  - Semantic relationships between metadata
- Heterogeneous data sources
  - Metadata integration
  - Heterogeneous data models
  - Monitoring, error recovery, fault tolerance

# 3. Knowledge Generation

- Gain knowledge/insight from peta-scale data
  - Data model specific characterizations that preserve information
  - Characterization of complexity of data model
- Data agents for feature detection
  - Dynamic extraction of knowledge, on-the-fly analysis
- Data provenance for tracking virtual data products
  - Tracking knowledge required to recreate data
  - Virtual data management reproducibility of data

# 3. Knowledge Generation

- Generic technology
  - Knowledge generation from petabytes of data
  - Quantification of insight
  - Qualitative organization of insight
  - Characterization of human visualization as a knowledge extraction process
  - Knowledge management

# 3. Knowledge Generation

- Characterization of problem space as knowledge generation
  - Application of knowledge generation rules to data on-the-fly
  - Analysis of knowledge for worthiness
- Characterization of visualizations as derived data products

# Visualization Challenges

- Data analysis guided visualization on-the-fly feature detection
- Exploration of large data sets without moving data to desk top
- Algorithms for latency management
- Algorithms for tuning presentation to level of resource through level of detail
- 3D and stereo visualization
  - Human interface
- Visualization of knowledge
- Cognitive human interface

#### Cross-Area Collaborations

- Operating System
  - Grid computing
    - Co-scheduling of CPU, storage, and I/O bandwidth
    - Distributed data management
      - replication vs Internet Backplane Protocol for usage-based data movement
    - Common data format
    - Fault tolerance
    - Bandwidth allocation
    - Global name space for persistent objects
    - Storage abstraction for manipulating data within storage

#### Cross Area Collaboration

- Portability / tools
  - Bandwidth parallel I/O support from application
  - Grid computing
    - Persistent objects
    - Common data format
- Performance
  - Bandwidth end-to-end
  - Data bottleneck identification
  - Network protocol support for QOS
- Programming models and run time
  - Visualization based simulation problem set-up
  - High-level data abstraction
  - Persistent parallel programming model

#### Coordination across Communities

- Unification Interoperability of technologies
- Unification of islands of expertise
  - Unification of approaches between data and vis
- Unification of application-drivers with computer science research

# Current Gaps in MICS CS Research

- Application of grid computing capabilities throughout MICS program
  - Intelligent data processing Data intensive architectures
    - Computation in data resource
    - Agent based systems for data access
  - Fault tolerant computing
- Integration of knowledge generation across data analysis and visualization communities
  - Knowledge extraction and visualization
  - Abstraction of complex data types
- Productization of software
  - porting to infrastructure independent representations (OMG-MDA)